Analysis of cancer patients useing Haberman Dataset

```
In [1]: | # Importing impotrtant library.
        import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        import numpy as np
        # Loadind haberman.csv Data to Dataframe.
        haberman= pd.read_csv("haberman.csv")
In [2]: | # Number of data point in the dataframe.
        print (haberman.shape)
        (306, 4)
In [3]: # Column name of the Dataset.
        print (haberman.columns)
        Index(['age', 'year', 'nodes', 'status'], dtype='object')
In [4]: # how many data poits from each of the class.
        haberman['status'].value counts()
        # this data set is not balance because the difference between data points in b
        oth the classes is very large.
Out[4]: 1
             225
              81
        Name: status, dtype: int64
```

About haberman dataset

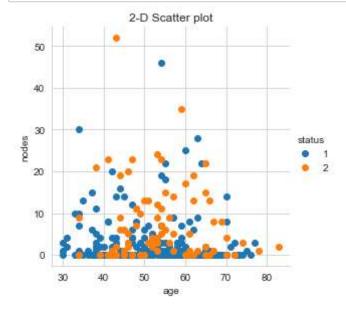
- this data set is about the cancer patients those who andergone through surgery between year of 1958 to 1969
- this data contains 4 attributs and 306 data points.
- · the four attributs are :-
 - 1. age = Age of patients
 - 2. year = In which year surery had been done.
 - 3. nodes = This is related to medical terminology, also known as Lymph node is an ovoid or kidney-shaped organ of thelymphatic system and the adaptive immune system. Lymph nodes are widely present throughout the body and are linked by the lymphatic vessels. in this data set it is of range of 0 to 52
 - 4. status: '1' is for survial of more than 5 year after surery, '2' is for survial of less than 5 year after surery.

Objective

 To make analogy betwen attributes and do analysis of the data points by ploting different types of plots and observe the plot and parametes like mean, median, std_deviation and make a observation about status (class) of the patients.

2-D Scatter plot

```
In [9]: sns.set_style('whitegrid');
    sns.FacetGrid(haberman, hue = 'status' , size = 4).map(plt.scatter , 'age', 'n
    odes').add_legend();
    plt.title('2-D Scatter plot')
    plt.show()
```



Obervation:-

- this is coloured scatter plot and the blue colour is of status of '1' and most of the blue one have lower nodes
- chances to survive patient is very high is age of patient is less then '60' and nodes is less than '5'
- but not much separation between status '1' and status '2'.

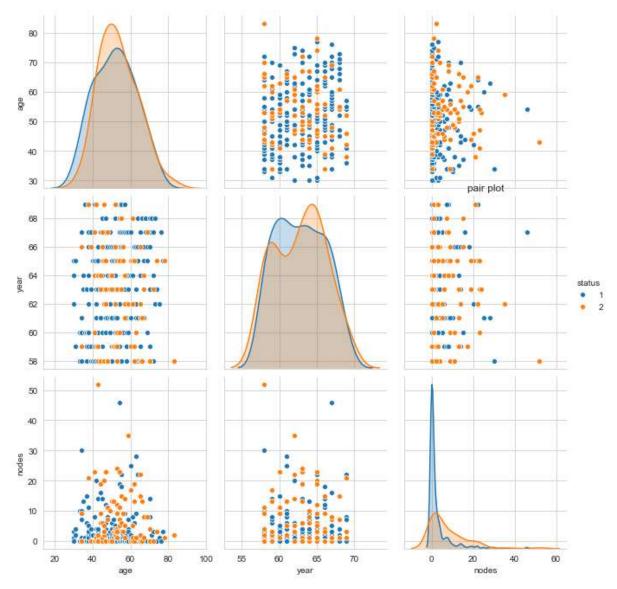
pair_plot

```
In [12]: # there are nC2 plot exis .
# the diagonal plot is PDF
#Basically this plot is to find satter plot which is best to analize , and you
you can also find which pair of data is best to
#analize
plt.close();
sns.set_style('whitegrid');
sns.pairplot(haberman , hue = 'status' , size = 3);
plt.title('pair plot')

plt.show()
```

C:\Users\anant\anaconda3\lib\site-packages\seaborn\axisgrid.py:2071: UserWarn
ing: The `size` parameter has been renamed to `height`; please update your co
de.

warnings.warn(msg, UserWarning)



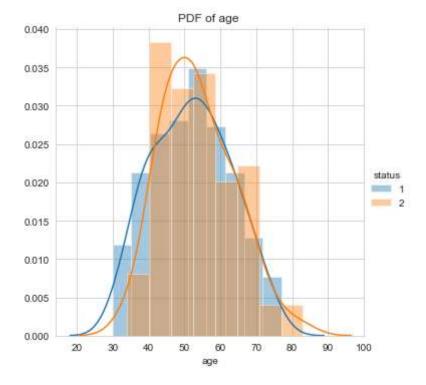
observation:-

• There is no such good result except plot between 'age' and 'nodes' which analysis have been done previously.

Histogram, PDF

C:\Users\anant\anaconda3\lib\site-packages\seaborn\axisgrid.py:243: UserWarni
ng: The `size` parameter has been renamed to `height`; please update your cod
e.

warnings.warn(msg, UserWarning)

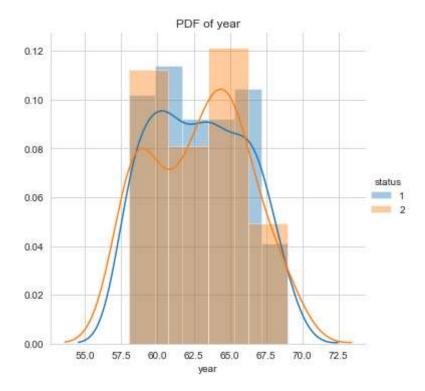


observation:-

• although a large portion is overlap but we can observe that patiens with age less then '40' have more chances to survive.

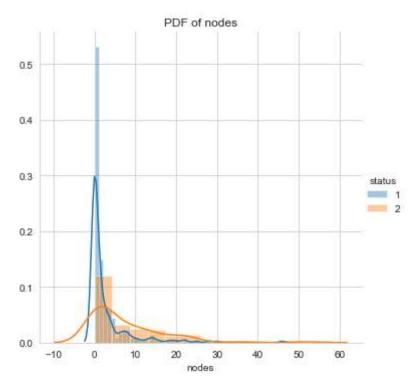
C:\Users\anant\anaconda3\lib\site-packages\seaborn\axisgrid.py:243: UserWarni
ng: The `size` parameter has been renamed to `height`; please update your cod
e.

warnings.warn(msg, UserWarning)



C:\Users\anant\anaconda3\lib\site-packages\seaborn\axisgrid.py:243: UserWarni
ng: The `size` parameter has been renamed to `height`; please update your cod
e.

warnings.warn(msg, UserWarning)



Obervation:-

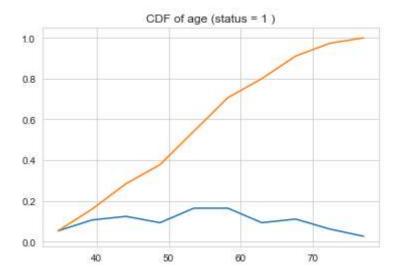
• nodes of value less than '5' have higher chances to survive.

CDF

```
In [19]: counts,bin_edges = np.histogram(haberman_1['age'],bins = 10,density = True)
    pdf = counts/(sum(counts))
    print(pdf)
    cdf = np.cumsum(pdf)
    plt.title('CDF of age (status = 1 )')
    plt.plot(bin_edges[1:],pdf)
    plt.plot(bin_edges[1:],cdf)
```

[0.05333333 0.10666667 0.12444444 0.09333333 0.16444444 0.16444444 0.09333333 0.11111111 0.06222222 0.02666667]

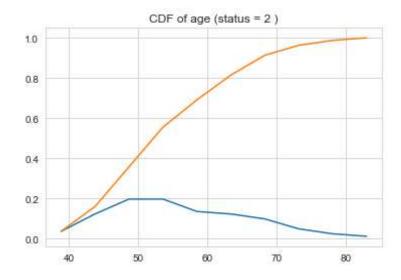
Out[19]: [<matplotlib.lines.Line2D at 0x173d287ebb0>]



```
In [20]: counts,bin_edges = np.histogram(haberman_2['age'],bins = 10,density = True)
    pdf = counts/(sum(counts))
    print(pdf)
    cdf = np.cumsum(pdf)
    plt.title('CDF of age (status = 2 )')
    plt.plot(bin_edges[1:],pdf)
    plt.plot(bin_edges[1:],cdf)
```

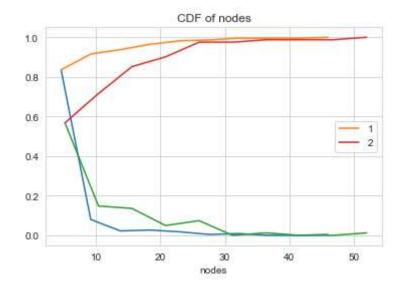
[0.03703704 0.12345679 0.19753086 0.19753086 0.13580247 0.12345679 0.09876543 0.04938272 0.02469136 0.01234568]

Out[20]: [<matplotlib.lines.Line2D at 0x173d2801550>]



```
In [21]:
         counts,bin_edges = np.histogram(haberman_1['nodes'],bins = 10,density = True)
         pdf = counts/(sum(counts))
         print(pdf)
         cdf = np.cumsum(pdf)
         plt.plot(bin_edges[1:],pdf)
         plt.plot(bin_edges[1:],cdf , label=1)
         counts,bin_edges = np.histogram(haberman_2['nodes'],bins = 10,density = True)
         pdf = counts/(sum(counts))
         print(pdf)
         cdf = np.cumsum(pdf)
         plt.plot(bin_edges[1:],pdf)
         plt.plot(bin_edges[1:],cdf ,label = 2)
         plt.xlabel("nodes")
         plt.legend()
         plt.title('CDF of nodes')
         plt.show()
```

```
[0.83555556 0.08 0.02222222 0.02666667 0.01777778 0.00444444 0.00888889 0. 0. 0.00444444] [0.56790123 0.14814815 0.13580247 0.04938272 0.07407407 0. 0.01234568 0. 0. 0.01234568]
```



```
In [69]: # Obsevation:-
# CDF started with more then 80% that means if nodes value is less more chanc
e to survive.
# if nodes value less then 5 then there is 90% chances to survive
```

Mean, and Std-dev

```
In [70]: print("Means nodes:")
         print(np.mean(haberman_1["nodes"]))
         print(np.mean(haberman_2["nodes"]))
         print("\nMean age \n:")
         print(np.mean(haberman_1["age"]))
         print(np.mean(haberman_2["age"]))
         print("\nStd-dev of nodes\n:");
         print(np.std(haberman_1["nodes"]))
         print(np.std(haberman_2["nodes"]))
         print("\nStd-dev of age:\n");
         print(np.std(haberman_1["age"]))
         print(np.std(haberman_2["age"]))
         Means nodes:
         2.791111111111113
         7.45679012345679
         Mean age
         52.017777777778
         53.67901234567901
         Std-dev of nodes
         5.857258449412131
         9.128776076761632
         Std-dev of age:
         10.98765547510051
```

Obsevation:-

- mean node value of patients who survive is 2.7 which is less and we have this in pervious analysis.
- from mean age we con not be able to any conclusion

10.10418219303131

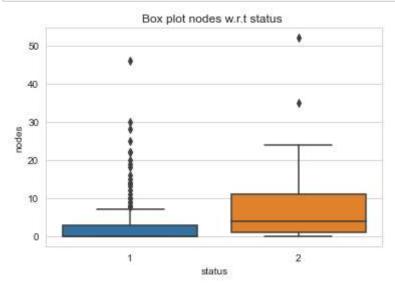
• spred of nodes of patientas who survive is less as compere to patients who had not survive.

Median, Percentile, Quantile, IQR, MAD

```
In [75]: | print("\nMedians:")
         print(np.median(haberman_1["nodes"]))
         print(np.median(haberman_2["nodes"]))
         print("\nQuantiles:")
         print(np.percentile(haberman["nodes"],np.arange(0, 100, 25)))
         print(np.percentile(haberman["nodes"],np.arange(0, 100, 25)))
         print("\n90th Percentiles:")
         print(np.percentile(haberman_1["nodes"],90))
         print(np.percentile(haberman_2["nodes"],90))
         from statsmodels import robust
         print ("\nMedian Absolute Deviation")
         print(robust.mad(haberman_1["nodes"]))
         print(robust.mad(haberman_2["nodes"]))
         Medians:
         0.0
         4.0
         Quantiles:
         [0. 0. 1. 4.]
         [0. 0. 1. 4.]
         90th Percentiles:
         8.0
         20.0
         Median Absolute Deviation
         0.0
         5.930408874022408
```

Box plot

```
In [27]: sns.boxplot(x='status',y='nodes', data=haberman )
   plt.title('Box plot nodes w.r.t status ')
   plt.show()
```

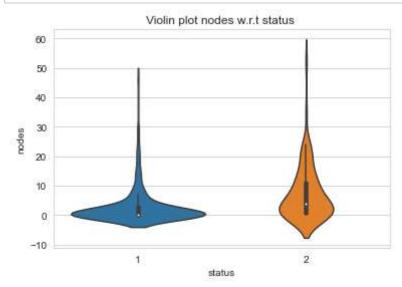


Observation:-

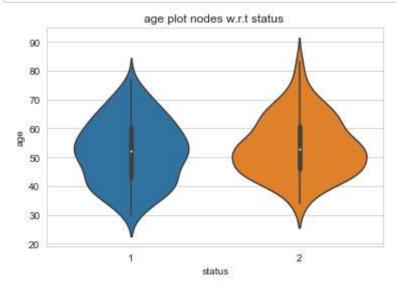
- from box plot also we can say that to most probably if nodes< "6" is to susrvive .
- if node value is more then '7' patient mosst probabily to died.

Violin plots

```
In [28]: # with nodes
sns.violinplot(x="status", y="nodes", data=haberman, size=10)
plt.title('Violin plot nodes w.r.t status ')
plt.show()
```



```
In [29]: # with age
    sns.violinplot(x="status", y="age", data=haberman, size=10)
    plt.title('age plot nodes w.r.t status ')
    plt.show()
```



Obervation:-

- by looking at both plot with 'nodes' and 'age' chances of survive a patients is high if age is between 40 to 60
- but at chances of die is almost same withe the age
- but in the plot with the node plot "1" is very much concentrated at "0" so if node value is less than chances of survive is high

summary

- In scatter plot numerous number of poits are closest to zero node axis and in nodes vs age plot so chances
 of survive is high three
- Mean of node value who survived is is 2.7 and mean who died is 7.4.
- Median of node value who servived is '0' and who died is '4'
- By looking at PDf there is high overlapping in every plot still be con con conclude that age is less than 40 and node value less than 5 then high chances of surivie.
- CDF started with more then 80% that means if nodes value is less more chance to survive. if nodes value less then 5 then there is 90% chances to survive.
- from box plot also we can say that to most probably if nodes< "6" is to susrvive . if node value is more then '7' patient mosst probabily to died.
- In the plot with the node is very much concentrated at "0" so if node value is less than chances of survive is high

```
In [ ]:
```

In []: